

PART 2

LM3-EUTRO

Chapter 5. Calibration

2.5.1 Description of Process

After model equations were formulated and coded as a computer program, model calibration was the next step. The goal of calibrating water quality models was to adjust the model coefficients in order to obtain the best possible fit between the model output and the field data. Challenges of calibrating eutrophication models included the many degrees of freedom (independent model coefficients) and the uncertainty of many of these model coefficients. A traditional model calibration approach was used for LM3-Eutro. The model coefficients were initially estimated using values and ranges reported in the literature (see Table 2.4.7) and these parameters were then adjusted to provide the best model fit to the field data. In this study, values for many coefficients were derived from available Lake Michigan and Great Lakes historical data collected by reputable agencies such as the United States Environmental Protection Agency's (USEPA) Great Lakes National Program Office (GLNPO), National Oceanic and Atmospheric Administration's (NOAA) Great Lakes Environmental Research Laboratory (GLERL) and the University of Michigan. Very few field and laboratory experiments were performed to estimate kinetic coefficients for LM3-Eutro. Limited ^{14}C primary production experiments were performed and used in determining phytoplankton growth coefficients. Phytoplankton (diatoms and non-diatoms), particulate organic carbon (POC), total phosphorus, and dissolved silica (DSi) were the most important state variables in model

calibration. However, all variables were evaluated during the calibration process.

The model was calibrated on the Level 3 segmentation framework (Figure 2.5.1). The main calibration emphasis was placed on the main lake due to inadequate Green Bay data. The high-resolution (Level 3) segments were also collapsed to the Level 2 segmentation scheme to provide a visual representation of how well the model reflected the field data in different areas of the lake and captured expected trends, such as spring epilimnetic diatom peaks and nutrient depletion. It was not

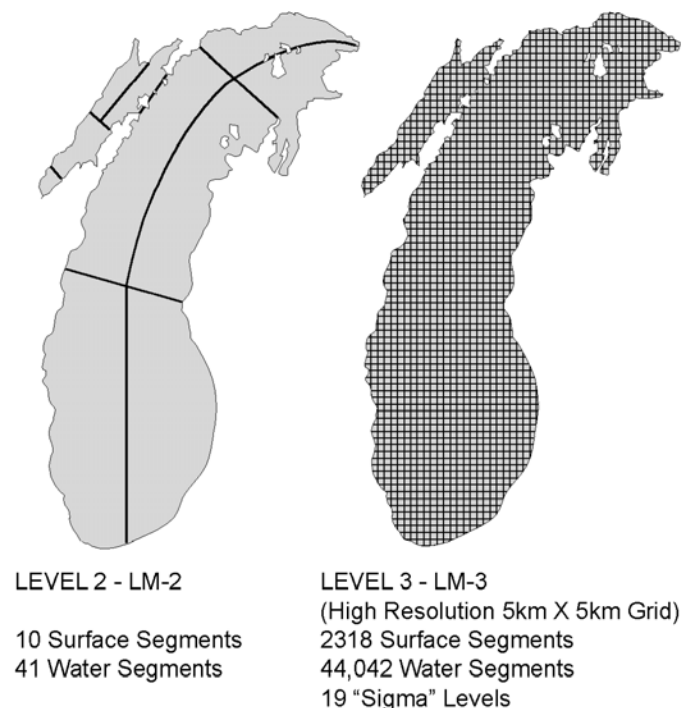


Figure 2.5.1. Level 2 and Level 3 model segmentation.

feasible to visually compare model output versus field data on the Level 3 segmentation framework due to the large number (44,042) of 5 km² cells. Instead, we regressed model output versus field data for each of the 5 km² cells where a field data point was available. This enabled calculation of simple statistical parameters such as square of the correlation coefficient (r^2) and slope and direct comparison of the different model calibration runs.

During the calibration process, small changes to the model initial conditions were made to observe the effect on the model fit. This was done because of the uncertainty of the initial condition estimates and the fact that the initial conditions have a significant influence on the model output.

Data from laboratory primary productivity experiments were used to constrain and confirm values for the growth coefficients that were used in the model. Productivity experiment results; light and temperature parameters; and the Lake Michigan Mass Balance Project (LMMBP) field data were applied to LM3-Eutro productivity equations to generate model production estimates comparable to those generated in the laboratory experiments. Model constants were then adjusted in order to best reflect the primary production trends observed in the laboratory experiments (Figure 2.5.2).

2.5.2 Selection of Best Calibration

After performing several hundred model simulations, we selected our best run based on statistical parameters including square of correlation coefficient (r^2) and slope. The coefficients of the final run were constrained to ensure that all model coefficients fell within reasonable and reported ranges. The best model fit was also evaluated visually on the Level 2 segmentation scheme. Important criteria included the model fit with the overall field data, ability to capture observed and expected phytoplankton peaks, and how well hypolimnetic and epilimnetic nutrient, carbon, and plankton trends and concentrations were predicted. The calibrated final model coefficients are listed in Table 2.5.1. Figure 2.5.3 shows the overall Level 3 fit for phytoplankton, total phosphorus, POC, and DSi. Statistical results are summarized in Table 2.5.2. Figure 2.5.4 shows model versus field data plots for selected Level 2 segments. Selected Level 3 5 km² cells representing

nearshore regions and offshore regions are presented in Figure 2.5.5. Model output for the 5 km² cells was much more dynamic than for the larger Level 2 segments and there were far fewer data points for model fit determination. The model appeared to fit the available data very well in some of the cells and not as well in others. Level 3 model versus data comparisons in individual 5 km² cells were not used in our calibration exercise.

Although we attempted to calibrate all of the state variables, less emphasis was placed on the nitrogen states because Lake Michigan is phosphorus and silica-limited. We also did not perform any comparison of model output with the field data for particulate silica (SU) or soluble reactive phosphorus (SRP), since there was no SU field data and more than 80% of the SRP field data fell below the detection limit. As stated, we did not spend much time calibrating Green Bay because of the limited sampling done in the bay. As a consequence, the final Green Bay calibration was not as good as the rest of the lake. This was especially true for the portion of the bay closest to the Fox River.

With the exception of zooplankton, the final model calibration was reasonably good, and the model was able to fit the field data well and capture important spatial and seasonal trends. A brief discussion of individual calibration results for phytoplankton, POC, total phosphorus, and DSi follows.

2.5.2.1 Phytoplankton

The model somewhat underestimated the field data for phytoplankton. Seabird chlorophyll *a* data were used for all phytoplankton field values. Part of the explanation for the underestimation was poor chlorophyll *a* field data. The Seabird fluorescence instrument, like many *in vivo* fluorescence methods, is notorious for its inaccuracy in measuring chlorophyll *a* (U.S. Environmental Protection Agency, 1997; Clesceri *et al.*, 1998). The square of the correlation coefficient of 0.37 was acceptable, especially given the inherent variation in phytoplankton communities over space and time. Our fit was in-range of other published eutrophication models (Thomann, 1982; Cerco and Cole, 1994). The model was able to capture spatial and temporal trends such as the spring diatom blooms (Figure 2.5.4) and earlier phytoplankton blooms in the

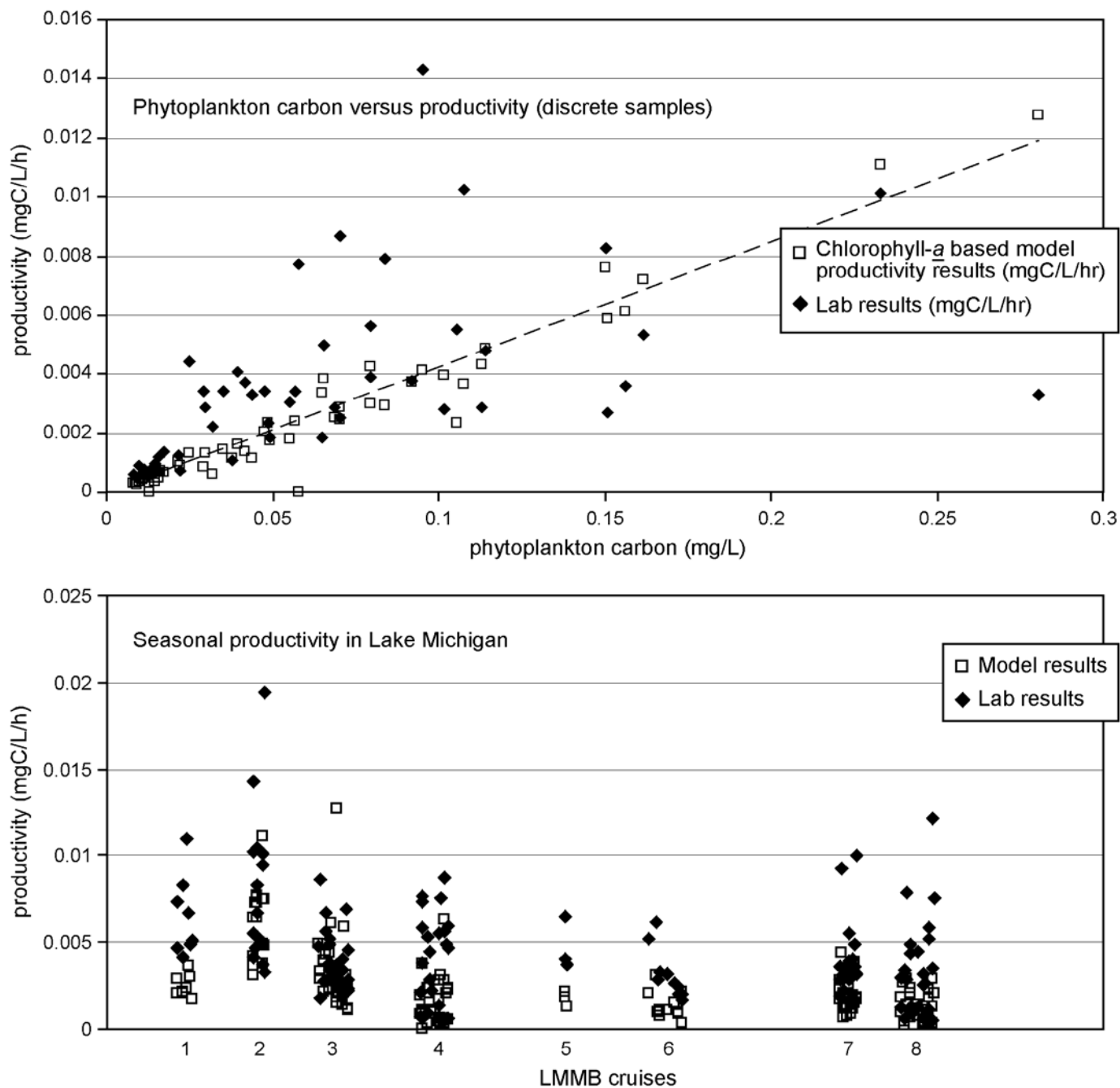


Figure 2.5.2. LM3-Eutro model versus laboratory primary production.

Table 2.5.1. Coefficients Used in the LM3 Model (Units Correspond to Required LM3 Model Output)

Coefficient	Value	Unit	Description
ANCP	0.25		Nitrogen:carbon ratio (mass basis)
APCP	0.01		Phosphorus:carbon ratio (mass basis)
ASCD	2.3		Silica:carbon ratio (mass basis)
BMRD	8.6E-07	1/s	Diatom mortality
BMRG	8.6E-07	1/s	Greens mortality
CCHLD	40		Carbon:chlorophyll ratio (diatoms)
CCHLG	40		Carbon:chlorophyll ratio (greens)
CGZ	3.1E-06	m ³ /kg/s	Zooplankton grazing rate coefficient
FCD	0.05		Dissolved organic carbon fraction from diatom mortality
FCDG	0.05		Dissolved organic carbon fraction from greens mortality
FCLD	0.3		Labile organic carbon fraction from diatom mortality
FCLG	0.3		Labile organic carbon fraction from greens mortality
FCRD	0.3		Refractory organic carbon fraction from diatom mortality
FCRG	0.3		Refractory organic carbon fraction from greens mortality
FCDP	0.35		Dissolved organic carbon fraction from algal predation
FCDZ	0		Dissolved organic carbon fraction from zooplankton mortality
FCLP	0.5		Labile particulate dissolved carbon fraction from algal predation
FCLZ	0.4		Labile particulate dissolved carbon fraction from zooplankton mortality
FCRP	0.15		Refractory particulate dissolved carbon fraction from algal predation
FCRZ	0.1		Refractory particulate dissolved carbon from zooplankton mortality
FNDD	0.5		Dissolved organic nitrogen from diatom mortality
FNDG	0.5		Dissolved organic nitrogen fraction from greens mortality
FNDP	0		Dissolved organic nitrogen fraction from algal predation
FNDZ	0		Dissolved organic nitrogen fraction from zooplankton mortality
FNID	0.5		Dissolved inorganic nitrogen fraction from diatom mortality
FNIG	0.5		Dissolved inorganic nitrogen fraction from greens mortality
FNIP	0.5		Dissolved inorganic nitrogen fraction from algal predation
FNIZ	0.5		Dissolved inorganic nitrogen fraction from zooplankton mortality
FNLD	0		Labile organic nitrogen fraction from diatom mortality
FNLG	0		Labile organic nitrogen fraction from greens mortality
FNLP	0.4		Labile organic nitrogen fraction from algal predation
FNLZ	0.4		Labile organic nitrogen fraction from zooplankton mortality
FNRD	0		Refractory organic nitrogen fraction from diatom mortality
FNRG	0		Refractory organic nitrogen fraction from greens mortality
RNRP	0.1		Refractory organic nitrogen fraction from algal predation
RNRZ	0.1		Refractory organic nitrogen fraction from zooplankton mortality
FPDD	0.1		Dissolved organic phosphorus fraction from diatom mortality
FPDG	0.1		Dissolved organic phosphorus fraction from greens mortality
FPDP	0.2		Dissolved organic phosphorus fraction from algal predation
FPDZ	0.2		Dissolved organic phosphorus fraction from zooplankton mortality
FPID	0.3		Dissolved inorganic phosphorus fraction from diatom mortality
FPIG	0.3		Dissolved inorganic phosphorus fraction from greens mortality
FPIP	0.5		Dissolved inorganic phosphorus fraction from algal predation
FPLD	0.5		Dissolved inorganic phosphorus fraction from zooplankton mortality
FPLG	0.3		Labile organic phosphorus fraction from greens mortality
FPLP	0.15		Labile organic phosphorus fraction from algal predation
FPLZ	0.15		Labile organic phosphorus fraction from zooplankton mortality
FPRD	0.3		Refractory organic phosphorus fraction from diatom mortality
FPRG	0.3		Refractory organic phosphorus fraction from greens mortality
FPRP	0.15		Refractory organic phosphorus fraction from algal predation
FPRZ	0.15		Refractory organic phosphorus fraction from zooplankton mortality

Table 2.5.1. Coefficients Used in the LM3 Model (Continued)

Coefficient	Value	Units	Description
FSAP	0		Dissolved silica fraction from diatom predation
GREFF	0.6		Zooplankton grazing coefficient
ILUMO	25	W/m ²	Constant illumination (first 90 days)
ISMIN	400	W/m ²	Optimum light illumination
KDC	1.16E-08	1/s	Dissolved organic carbon mineralization coefficient
KDCALG	0.00E+00	m ³ /kg/s	Dissolved organic carbon algal dependency coefficient
KDN	1.74E-07	1/s	Dissolved organic nitrogen mineralization coefficient
KDNALG	0.00E+00	m ³ /kg/s	Dissolved organic nitrogen algal dependency coefficient
KDP	1.16E-09	1/s	Dissolved organic phosphorus mineralization coefficient
KDPALG	6.0E-03	m ³ /kg/s	Dissolved organic phosphorus algal dependency coefficient
KE	0.15	1/m	Background light attenuation
KECHL	1.7E+04	m ² /kg	Light attenuation for chlorophyll <i>a</i>
KHND	2.50E-05	kg/m ³	Nitrogen half-saturation coefficients for diatoms
KHNG	2.50E-05	kg/m ³	Nitrogen half-saturation coefficients for greens
KHNNT	0.0000	kg/m ³	Nitrate half-saturation coefficient for nitrification
KHPD	5.0E-07	kg/m ³	Phosphorus half-saturation coefficients for diatoms
KHPG	5.0E-07	kg/m ³	Phosphorus half-saturation coefficients for greens
KHSD	6.0E-05	kg/m ³	Silica half-saturation coefficients for diatoms
KLC	1.0E-07	1/s	Labile organic carbon hydrolysis coefficient
KLCALG	0.00E+00	m ³ /kg/s	Labile organic carbon algal dependency coefficient
KLN	3.47E-07	1/s	Labile organic carbon hydrolysis coefficient
KLNALG	0.00E+00	m ³ /kg/s	Labile organic nitrogen algal dependency coefficient
KLP	1.00E-09	1/s	Labile organic phosphorus hydrolysis coefficient
KLPALG	6.00E-03	m ³ /kg/s	Labile organic phosphorus algal dependency coefficient
KRC	1.00E--07	1/s	Refractory organic carbon hydrolysis coefficient
KRCALG	0.00E+00	m ³ /kg/s	Refractory organic carbon algal dependency coefficient
KRN	3.47E-08	1/s	Refractory organic nitrogen hydrolysis coefficient
KRNALG	00E+00	m ³ /kg/s	Refractory organic nitrogen algal dependency coefficient
KRP	1.0E-09	1/s	Refractory organic phosphorus hydrolysis coefficient
KRPALG	6.0E+03	m ³ /kg/s	Refractory organic phosphorus algal dependency coefficient
KSUA	2.5E-07	1/s	Biogenic silica dissolution rate
KSZ	1.0E-04	kg/m ³	Zooplankton half-saturation (for algae)
KTBD	0.074	1/°C	Diatom mortality temperature coefficient
KTBG	0.074	1/°C	Greens mortality temperature coefficient
KTGD1	0.0025	1/°C ²	Diatom growth temperature coefficient (< optimum)
KTGD2	0.006	1/°C ²	Diatom growth temperature coefficient (> optimum)
KTGG1	0.0025	1/°C ²	Greens growth temperature coefficient (< optimum)
KTGG2	0.006	1/°C ²	Greens growth temperature coefficient (> optimum)
KTHDR	9.9E-02	1/°C	Hydrolysis temperature dependency coefficient
KTMNL	7.4E-02	1/°C	Mineralization temperature dependency coefficient
KTNT1	0.004	1/°C ²	Nitrification temperature coefficient (< optimum)
NTNT2	0.004	1/°C ²	Nitrification temperature coefficient (> optimum)
KTSUA	0.069	1/°C	Silica dissolution temperature coefficient
NTM	0.074	1/°C	Diatom mortality temperature coefficient
NTM	2.50E-11	kg/m ³ /s	Nitrification rate coefficient
PMD	2.90E-05	1/s	Diatom growth coefficient
PMG	2.60E-05	1/s	Greens growth coefficient
TMD	18	°C	Optimum diatom growth temperature
TMG	18	°C	Optimum greens growth temperature
TMNT	30	°C	Optimum nitrification temperature
TRD	20	°C	Optimum diatom mortality temperature

Table 2.5.1. Coefficients Used in the LM3 Model (Continued)

Coefficient	Value	Units	Description
TRG	20	°C	Optimum greens mortality temperature
TRHDR	20	°C	Optimum hydrolysis temperature
TRMNL	20	°C	Optimum mineralization temperature
TRSUA	20	°C	Optimum silica dissolution temperature
TZREF	20	°C	Optimum predation temperature
ZDTH	5.0E-07	1/s	Zooplankton mortality rate coefficient
ZTHET	1.0		Arrhenius temperature coefficient for predation
VDIA	1.15E-06	m/s	Diatoms settling coefficient
VGRE	0.85E-06	m/s	Greens settling coefficient
VLOC	2.0E-06	m/s	Labile organic carbon settling coefficient
VROC	2.0E-06	m/s	Refractory organic carbon settling coefficient
VLON	2.0E-06	m/s	Labile organic nitrogen settling coefficient
VRON	2.0E-06	m/s	Refractory organic nitrogen settling coefficient
VLOP	2.0E-06	m/s	Labile organic phosphorus settling coefficient
VROP	2.0E-06	m/s	Refractory organic phosphorus settling coefficient
VSU	2.0E-06	m/s	Biogenic silica settling coefficient

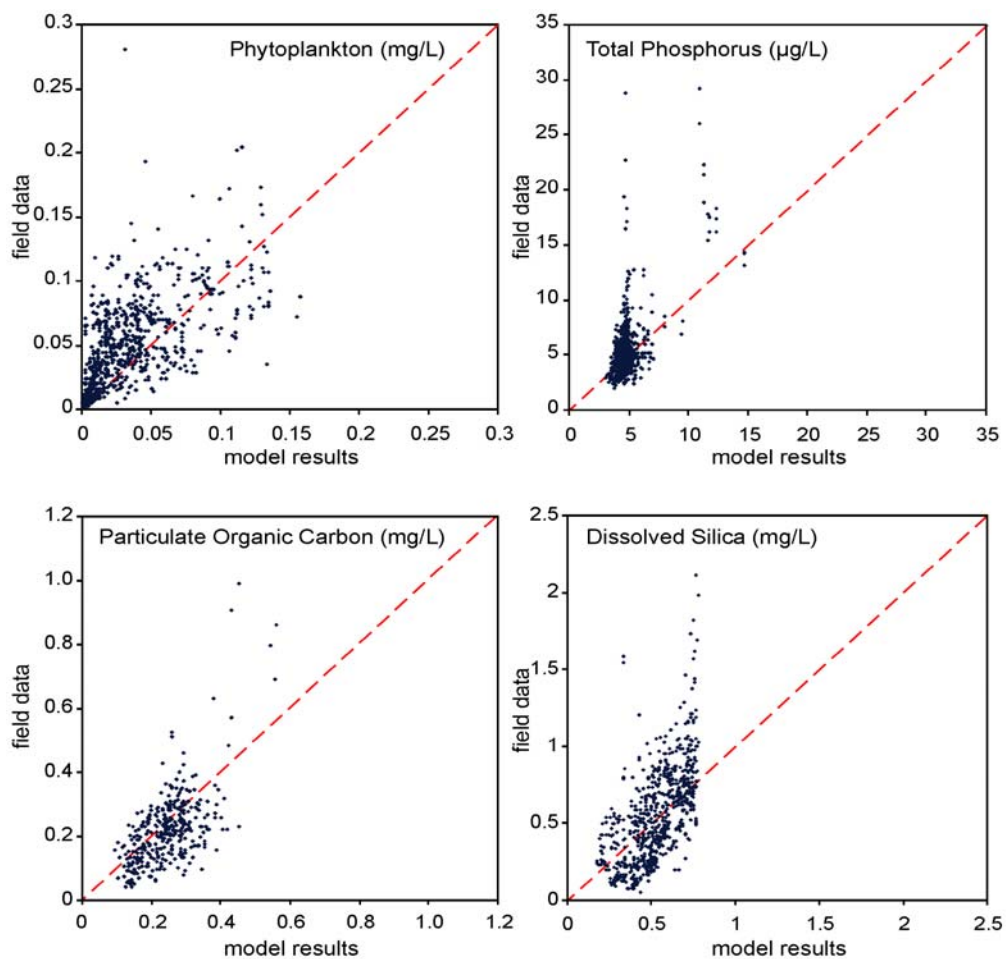


Figure 2.5.3. Level 3 LM3-Eutro model predictions versus field data, lake-wide.

Table 2.5.2. Summary of Statistical Results of the Calibration

Variable	Regression Coefficient (r^2)	Slope
Phytoplankton	0.37	0.67
Particulate Organic Carbon	0.39	0.95
Total Phosphorus	0.37	1.4
Dissolved Silica	0.37	1.2
Zooplankton	0.13	0.43

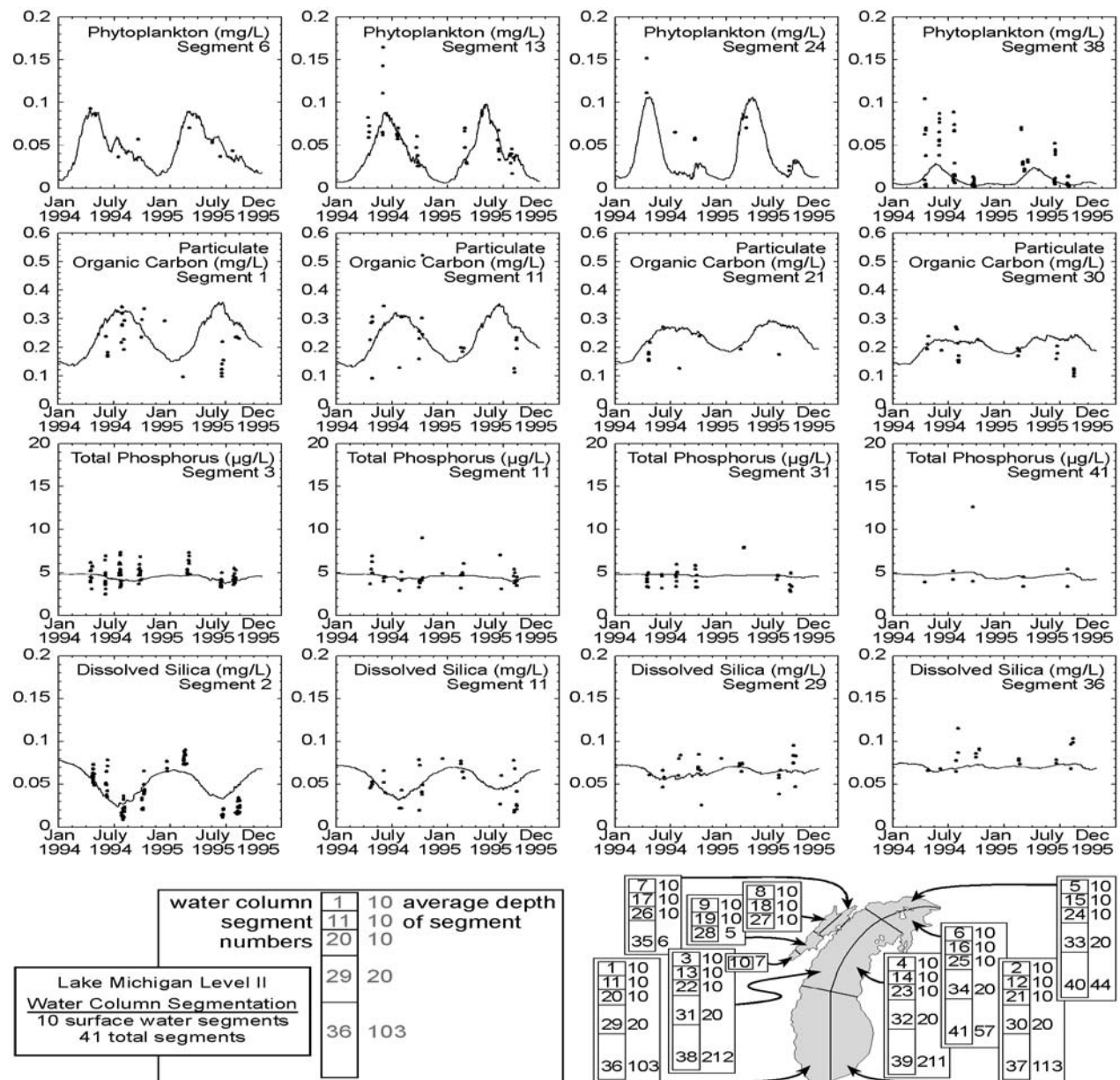


Figure 2.5.4. Level 2 LM3-Eutro model output versus field data for selected segments.

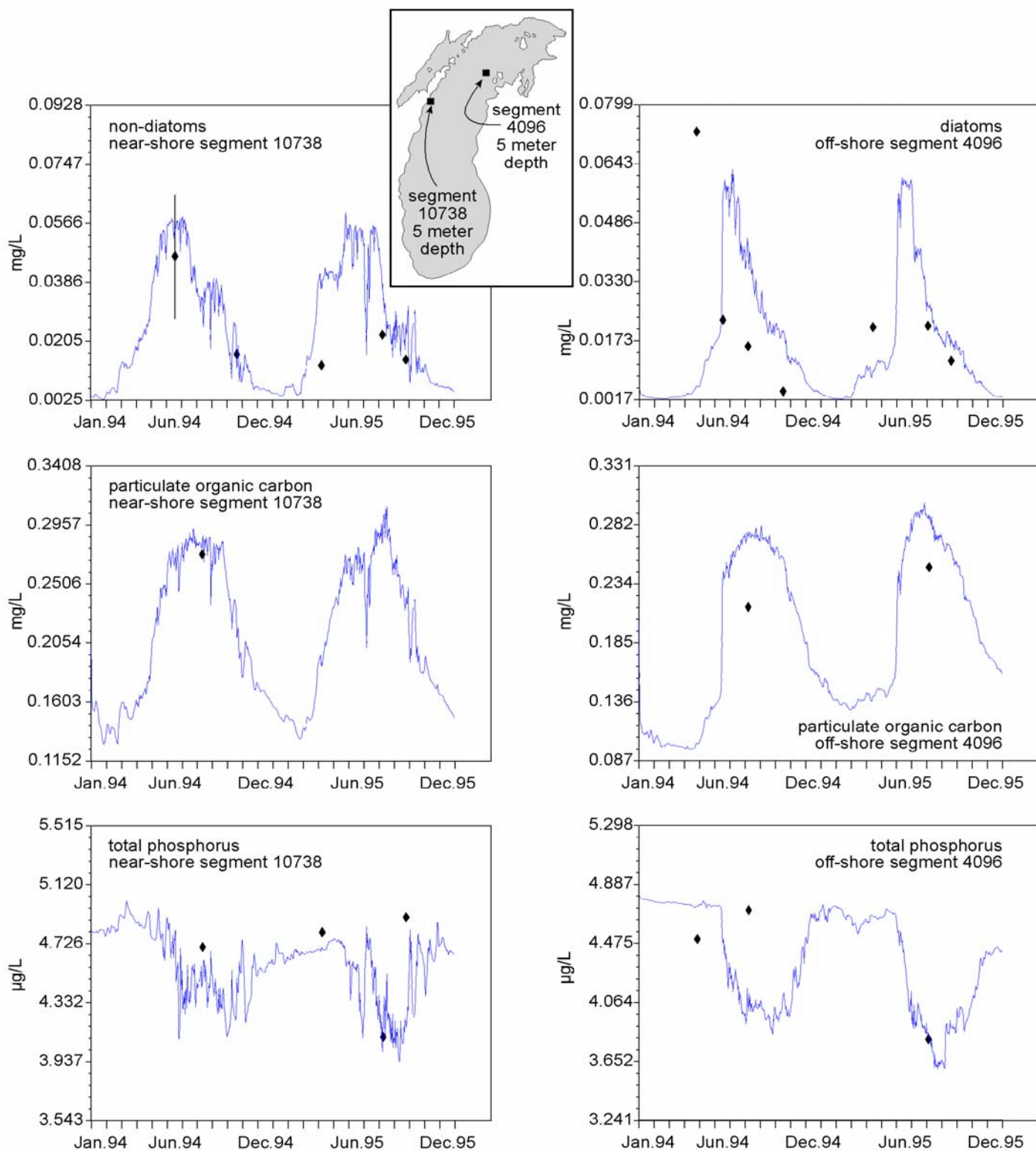


Figure 2.5.5. Level 3 LM3-Eutro model output versus field data for selected nearshore and offshore cells.

nearshore cells than in the offshore cells (Figure 2.5.5).

2.5.2.2 Particulate Organic Carbon

The model fits the field data well (Figure 2.5.3), with a slope of almost one (0.95) and a square of the correlation coefficient of almost 0.4 ($r^2 = 0.39$). The POC data exhibited less scatter than the other variables and were reflective of the phytoplankton in the lake. Additionally, we had a great deal of confidence in the POC data measurement technique. The ability of the model to capture the POC data trend increased our confidence in the model's overall eutrophication predictions. Examination of Level 2 segments showed that the model captured important trends, such as higher POC concentrations during the spring diatom bloom (Figure 2.5.4).

2.5.2.3 Total Phosphorus

The model fits the total phosphorus data reasonably well with a slope of 1.4 and a square of the correlation coefficient of 0.37. In general, the total phosphorus concentrations in the lake were fairly constant, with most measurements falling between 4 and 5 $\mu\text{g/L}$ and little seasonal variation observed. The overall model fit was acceptable (Figure 2.5.4). However, there were several higher total phosphorus values measured in Green Bay close to the Fox River and at other nearshore stations close to rivers and/or areas where there were significant sediment resuspension. In these cases, the model was not able to mimic the data, probably due to initial conditions that were too low and the lack of a sediment resuspension term.

2.5.2.4 Dissolved Silica

The dissolved silica model fits the field data well, with a slope of 1.2 (influenced by several very high silica field data points) and a regression coefficient of 0.37.

The model predicted the expected trends, with highest silica in the winter, a steep decline in the epilimnion in the spring coinciding with the diatom bloom, and a recovery toward the end of the year (Figure 2.5.4).

In general, the model fit the field data reasonably well. Seasonal and spatial trends for important variables, such as phytoplankton and silica, were captured. There were difficulties in accurately predicting Green Bay phytoplankton and nutrient concentrations, but this was not unexpected.

References

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